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EXHIBIT B

LibGen dataset: 650B* clean & deduped tokens

POC: Nikolay Bashlykov

TL;DR: We have collected a new **650B*** dataset of high-quality tokens on almost every possible subject from STEM and fiction books to cooking, gardening and historic books.

*using GPT-4 tokenizer

Note: https://fb.workplace.com/

Slides: Fair-Use Lib 230713

Description:

- Library Genesis, or LibGen, is a search engine and digital library that provides free access to a vast collection of books, articles, and other scholarly materials. It was established as a response to the limited access and high costs of academic publications, aiming to make knowledge more widely available.
- LibGen's database includes content from a wide range of disciplines, including science, technology, engineering, mathematics (STEM), humanities, and social
 sciences. The platform offers PDF and EPUB (ZIP archive containing a collection of HTML, CSS, ...) versions of books and articles, often sourced from
 copyrighted materials without the permission of the copyright holders.
- There are three main collections in LibGen:
 - fiction spans 2.7 million fiction books, 5.6TB
 - sci-tech spans 3.7 million scientific books, 59.4TB
 - sci-mag spans 81 million scientific articles, 80.6TB
 - [TBD] there is also comics, 94.5TB
- Analogues:
 - Sci-Hub: similar to the sci-mag part of LibGen.



- Z-lib: initially a mirror of LibGen, but then evolved to a separate project. Now claims to have 23M books and 285B articles. Banned multiple times, but seems to be working currently. Worth investigating.

The PDFs are parsed with the NOUGAT library

LibGen	(full DB)	fiction	sci-tech	sci-r	nag Total
Total documents (#	*)	2,693,056	3,706,772	81,903,	411
Unique documents	(author&title)	1,607,593	3,274,071	72,624,	976
Language (%)		English: 65% German: 11% French: 6%	English: 51% Russian 29% German: 5%	N/A	
Format (%)		Epub: 59% PDF: 11% mobi: 10%	Epub: 16% PDF: 65% djvu: 11%	PDF: ~1	00%
Median number of	pages per doc (#)	170	258	6	
Extracted EN clear	tokens (#)	110B	220B	325B	
Deduped EN tokens (gpt-4 tokenizer)		70B	190B	320B	
Extracted non-EN	clean tokens (#)	55B	15B		
Extracted ALL clean&deduped tokens (#)	125B	205B	320B	650	0B

Libgen Part (pdf/epub/mobi)	Total (doc num)	Downloaded (doc num / %)	Parsed (doc num / %)	Location Raw	Location Processed	Location minhash deduped	Cleaned tokens (#)
Sci-tech EN	1,726,719 (454,064 epubs + 1,272,655 pdfs)	1,695,684 / 98%	1,496,473 / 88%	fair- use/scitech/	/fair- use/scitech/processed/en/2 0230526/ /fair_llm lbgen/scitech/s	minhashdeduped/lib/scitech/20 23112U/ fair_llm/ en/scitech/scitech_en_2023112U fair_llm_v3. minhas	220B ->190B deduped
Fiction EN	1,159,720 (1,041,740 epubs + 117,980 pdfs)	1,138,296 / 98%	1,042,125 / 92%	air- use/fiction/	data/data_v3/fair- use/fiction/processed/en/2 0230526/ fair_lln shuffled/libgen/fiction/fic tion_en	Fiction safe (w/o adult content) minhashdeduped/lib/fiction/202 31210/safe Fiction rest (w adult content) tlata/minhashdeduped/lib/fiction/202 31210/rest /fair_llm_v3/c hdeduped/lib/fiction-20231210	110B -> 70B deduped
Sci-mag EN	81,903,411 (876 chunks both EN and non-EN, but we can parse only EN)	847 / 96%	54.7M / 67%	data/fair- use/scimag/	fair- use/scimag/processed/en/2 0230726 fair_llm/ libgen/scimag/	minhashdeduped/lib/scimag/202 31120/ air_llm/libgen/s cmrag/scimag_20231120 /fair_llm_v3/minhas hdeduped/lib/scimag-20231120	3258 -> 3208 deduped

Libgen Part (pdf/epub/mobi)	Total (doc num)	Downloaded (doc num / %)	Parsed (doc num / %)	Location Raw	Location Processed	Location minhash deduped	Cleaned tokens (#)
Sci-tech non-EN	130,593 (123,281 epub + 7,312 mobi)	128,722 / 99%	118,589 / 92%	hir- use/scitech/epub_ non_en/	pata/data_v3/fair- use/scitech/processed/non _en/20231126/		15B
					ibgen/scitech/s citech_non_en_20231126		
					/fair_llm_v 3/gata/gata_vs/fair- use/scitech/processed/non _en/scitech-20231126		
Fiction non-EN	594,348 (545,578 epub + 48,770 mobi)	586,240 / 99%	461,246 / 79%	fair- use/fiction/epub_n on_en	air- use/fiction/processed/non_ en/20231126/		55B
					fair_llm, libgen/fiction/fic tion_non_en_20231126		
				P	fair_llm_v 3/data/data_v3/fair- use/fiction/processed/non_ en/fiction-20231126		
Total (gpt-4 tokenizer)							725B -> 650B

Libgen Part (pdf/epub/mobi)	Total (doc num)	Downloaded (doc num / %)	Parsed (doc num / %)	Location Raw	Location Processed	Location minhash deduped	Cleaned tokens (#)
							deduped

SUMMARY TABLE

Updates:

26.11.2023

Multilingual LibGen v2

Similar cleaning steps were applied to multilingual libgen (fiction and scitech) as well, except for token distribution KL divergence heuristics,

- We did not apply the token distribution outliers heuristics because the top documents returned by high KL divergence do not show clear patterns of repetition or
 ungrammatical text in multilingual libgen. Part of the reason is that we concatenated all non-English documents together, so the corpus is not homogenous for the tool to
 be useful. We decided to skip this step for multilingual in the short term, and we can revisit it later when we split the data by language.
- Overall, we removed 1% and 0.67% of total characters from fiction and scitech respectively. Impact from specific filters are included below.

	Fiction	Scitech
REPETITION	520	242
PII	31770	21233
Copyright	52264	30304
Excessive new line characters removed	1097379580	202740904

Location (RSC)			
o fiction:	fair_llm	libgen/	
 scitech 	fair_IIm	libgen	

- Examples of filtered data
 - Repetition

REPETITION: Fanculo! Grazie! Fanculo! Gr

PII : Als Paula erschöpft und müde nach der immerhin effektiven Zeit mit Annemarie noch ihre E-Mails abrief, um zu sc hauen, ob weitere Infos vom Prüfungsamt gekommen waren, stölperte sie über eine Mail von Sven349@gmx.de. Ohne Betreff. Aufgeregt klick te sie sie an.

: Ihnen, liebe Leserinnen und Leser, danke ich, dass Sie dieses Selfpublishing-Projekt durch den Kauf und das Les en unterstützt haben und würde mich sehr über Ihre Rückmeldung auf Amazon oder unter kontakt@cappuccino-romane.de freuen. Wenn Sie übe r Neuerscheinungen unterrichtet werden machten, schicke

o Copyri	ght						
Copyright:	ISBN	978	90	414	1567	7	
Copyright;	ISBN	978	90	414	1789	3	
Copyright	ISBN	978	90	414	1534	9	
Copyright:	ISBN	978	90	414	1684	1	

For intermediate output (e.g. what's being filtered), check out the following directories:

Scitech: fair_llm/ bgen
 Fiction: /fair_llm/ ibgen

17.11.2023

LibGen v2

There are a few improvements we can make to LibGen after the manual inspection of the datasets:

- Remove documents, highlighted by the Token Distribution tool: Token Distribution of Training Datasets
- Remove excessive new line character "\n\n\n\n":
 - Limit all the new line characters to 1 "\n"
- Remove repetition:
 - Remove lines that contain <8% unique words, but with at least 100 words
- Remove emails (PII data):

mail 5-ans - Lettermbile (1,71) 20 - 32 - 42 - 42 - 1,710 (1.11)

- Remove rows containing copyright in the first and last 25% of the book:
- Rows containing any of these words: ["ISBN", "Copyright", "©", "All rights reserved", "DOI"]
- [not used] Remove tables of Contents / References / Acknowledgements in the end of the book
 - Remove all rows after these words if happen in the last 25% of the document: ["Content", "References", 'About the author", "Acknowledgements"]
 - Remove rows with "Content" if happen in the first 10% of the document until the first row that has length more than 30 characters
- [TBD] Split content to Adult/General for LibGen Fiction

Implementation: https://github.com/fairinternal

Commented [1]: any rationale of why we're doing this? just better knowledge density? I wonder if it could be useful for long-context?

More details:

- Observations on LibGen-SciMag
- Data Review: libgen-fiction-books

What was filtered?

We filtered data inside of the documents as well as full documents (based on the Token Distribution outliers):

- Scitech: 0.85%
- Scimag: 0.28%
- Fiction: 1.17%
- scitech: total number of docs: 1255945 | {lines_copyright_removed': 2334655, 'newlines_removed': 2957148318, 'lines_pii_removed': 1808248, 'lines_repetition_removed': 190613}
- scimag: total number of docs: 41767181 | ('lines_copyright_removed': 16394972, 'newlines_removed': 4191208457, 'lines_pii_removed': 15212651, 'lines_repetition_removed': 410558}
- fiction: total number of docs: 760097 | {'lines_copyright_removed': 125855, 'newlines_removed': 1695675744, 'lines_pii_removed': 101729, 'lines_repetition_removed': 2448}

Copyright&PII (rows removed inside the documents)

Meta_Kadrey_00065251

Copyright: Copyright © Adeline Catherine Anderson, 2009
PII: Harper loves hearing from readers and if you'd like to drop her a note you can do so via
harperbliss@gmail.com
PII: Email me at cassandradee.author@gmail.com with questions and comments.
PII: Did you enjoy this book? We love to hear from our readers. Please email us at readerfeedb
ack@titanemail.com or write to us at Reader Feedback at the above address.
PII: **readerfeedback@titanemail.com**
PII: Thank you for reading. If you enjoyed this book, please leave a review . If you'd like to
send along private feedback or join my ARC team to get free Advanced Review Copies of my books, please email me at authorjamieknight@gmail.com
PII : e-mail: happywuyuandi@163.com
PII: e-mail: wnh@mail.nefu.edu.cn Copyright: Mobile GIS; Mobile Agent; Forest intelligent administration system; wireless communication
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ersity, Jeddah, Saudi Arabia under Grant 5-135-36-RG.Z. Li and M. Shahidehpour are with the Giavin Center for El ectricity Innovation, Illinois Institute of Technology, Chicago, IL 60616 USA (e-mail: zhiyi.li@hawCopyright: * [8] Z. Li and M. Shahidehpour, "Bilevel model for analyzing coordinated cyber-physical at tacks on power systems," _IEEE Trans. Smart Grid_, available online. DOI: 10.1109/TSG.2015.2456107.
Copyright: # COPYRIGHT
Copyright: They cannot be sold, shared or given away as it is an infringement on the copyright of this work: Her muse, a cross between Jimmy Stewart and Hugh Jackman, brings her stories to life for her readers in a way that the thing the them coming back time and again for more. Her favorite genre is paranormal romance with a great deal of spice. You can visit Kathi
online and drop her an email if you'd like. She lo
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PII: Reading, writing, and white-water rafting are the three things she enjoys the most. You can visit her at www.Anit raMcLeod.com, write to her at alm@AnitraMcLeod.com, or fan her at www.facebook.com/pages/Anitra-Lynn-McLeodCopyright: Copyright 1987 by Dale Brown.

Repetition (Caused by PDF parsing OCR model hallucination. Also removed inside the documents)

REPETITION____: CZ GORTON,1 M PAJO,1 KA RONLUND,2 DB RUSSELL,1 CS SENDALL2\({}^{\it{1}}\)Sexual Health Se rvice, Cairns Base Hospital, Cairns, Queensland, Australia \({}^{\it{2}}\)Department of Gastroenterology, Cairns Base Hospital, Cairns, Queensland, Australia \({}^{\it{4}}\)Department of Gastroenterology, Cairns Base Hospital, Cairns, Queensland, Australia \({}^{\it{4}}\)Department of Gastroenterology, Cairns Base Hospital, Cairns, Queensland, Australia \({}^{\it{5}}\)Exval Health Service, Cairns Base Hospital, Cairns, Queensland, Australia \({}^{\it{6}}\)Exval Health Service, Cairns Base Hospital, Cairns, Queensland, Australia \({}^{\it{7}}\)Exval Health Service, Cairns Base Hospital, Cairns, Queensland, Australia \({}^{\it{9}}\)Exval Health Service, Cairns Base Hospital, Cairns, Queensland, Australia \({}^{\it{10}}\)Exval Health Service, Cairns Base Hospital, Cairns, Queensland, Australia \({}^{\it{10}}\)Exval Health Service, Cairns Base Hospital, Cairns, Queensland, Australia \({}^{\it{11}}\)Exval Health Service, Cairns Base Hospital, Cairns, Queensland, Australia \({}^{\it{11}}\)Exval Health Service, Cairns Base Hospital, Cairns, Queensland, Australia \({}^{\it{11}}\)

REPETITION : The smell of the fly-a-later doing its wonderful work on those tasty french tries. This s mell was not always present, but children have been exposed to this small enough that they are able to calides, just as was too old to get excited conjure it in their entorhinal cortex whencerer they see a McDonald's. The sm ell itself had been carefully designed and artificially synthesized in a smell and taste (actory on the haden's been used). This was not always present, but children have been exposed to this small enough that they are able to calides, just as was too old to get excited conjure it in their entorhinal cortex whencerer they see a McDonald's. The smell itself had been carefully designed and artificially synthesized in a smell and taste (actory on the haden's been used). This was not always present, but children have been used to calides, just as was too old to get excited conjure it in their entorhinal cortex whencerer they see a McDonald's. The smell itself had been carefully designed and artificially synthesized in a smell and taste (actory on the haden's been used). This was not always present, but children have been used to calides, just as was too old to get excited conjure it in their entorhinal cortex whencerer they see a McDonald's. The smell itself had been carefully designed and artificially synthesized in a smell and taste (actory on the haden's been used). This was not always present, but child ren have been used to calides, just as was too old to get excited conjure it in their entorhinal cortex whencerer they see a McDonald's. The smell itself had been carefully designed and artificially synthesized in a smell an asmell and artificially synthesized in a smell itself had been carefully designed and artificially synthesized in a smell an asmell and artificially synthesized in a smell an asmell and artificially synthesized in a smell and artificially synthesized i

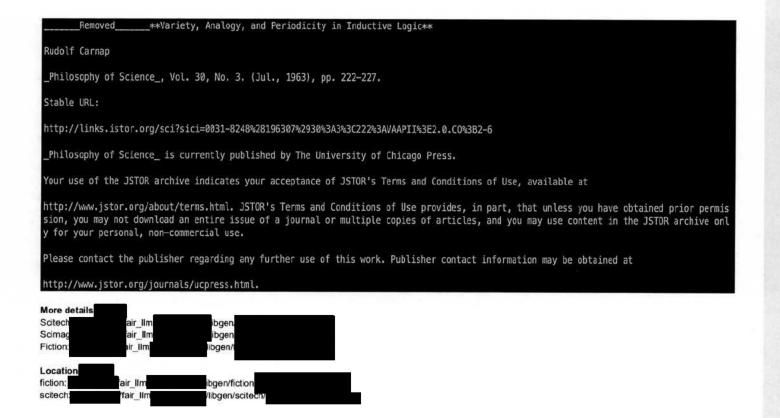
Meta_Kadrey_00065253

REPETITION_____: The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE. The Bitcoin PRACTICE is a very important part of the Bitcoin PRACTICE.

Removed documents (0.25% outliers based on Token Distribution. Removed full documents):

```
_Removed_____Sonata No. 1 in C Major Op. 1.
Sonata No. 1 in C Major Op. 1.
Sonata No. 1 in C Major Op. 1.
## References
* [1]
Figure 1: _A simple example of a \(p\)-component model._Sonata No. 1 in C Major Op. 1
The small notes may be omitted if necessary.
Sonata No. 1 in C Major Op. 1Sonata No. 1 in C Major Op. 1Sonata No. 1 in C Major Op. 1Sonata No. 1 in C Major O
p. 1.
Sonata No. 1 in C Major Op. 15onata No. 2 in F# Minor Op. 25onata No. 2 in F# Minor Op. 25onata No. 2 in F# Mino
r Op. 2Sonata No. 2 in F# Minor Op. 2Sonata No. 2 in F# Minor Op. 2Sonata No. 2 in F# Minor Op. 2Sonata No. 2 in
F# Minor Op. 2Sonata No. 2 in F# Minor Op. 2Sonata No. 2 in F# Minor Op. 2Sonata No. 2 in F# Minor Op. 2Sonata
No. 2 in F# Minor Op. 2.
Sonata No. 2 in F# Minor Op. 2Sonata No. 2 in F# Minor Op. 2.
Sonata No. 2 in F# Minor Op. 2.
```

Meta_Kadrey_00065255



HIGHLY CONFIDENTIAL - SOURCE CODE



Results:

The new mix shows improvement on most of the benchmarks. Low result on mmlu could be explained by high volatility of this benchmark (for example, on step 40k the result is 26.62, which is 1.6 points higher then on step 50k)

Caveat: we compare the results for step 42.5k, since at the moment we didn't have more GPUs to complete the training. The "7B Llama2 + LibGen-v1" is the most relevant baseline, as the difference is the version of LibGen + Open Web Math.

Step 42.5k	7B Llama2 + LibGen v2 + OWM (step 42.5k)	7B Liama2 Dill (step 42.5k)	7B Llama2 + Libgen-v1 (step 42k)	the same of the sa	Delta vs Llama2 Cin + LlbGen-v1
hellaswag.0_shot.acc_char	69.85	68.79	67.65	1.06	2.20
math.4_shot.1_gen.em	1.30	1.68		-0.38	n/a
nq.5_shot.em	17.65	17.40	13.38	0.25	4.27
tqa,5_shot.em	43.78	43.58	40.24	0.20	3.54
piqa.0_shot.acc_char	76.66	76.55	75.41	0.11	1,25
siqa.0_shot.acc_char	47.03	46.21	45.80	0.82	1.23
mmlu.5_shot.macro_avg.acc_char	24.05	24.14	25.96	-0.09	-1.91
human_eval.0_shot.1_gen.em	2.44	1.83	1.83	0.61	0.61
arc_challenge.0_shot.acc_char	40.34	40.26	38.28	0.09	2.06
ppl.code_py	4.06		4.44	n/a	0.39

Step 50k	14 C TOO 17 CO	7B Llama2 + Libgen-v1 (step 48k)	7B Llama2 + Libgen-v1 (step 51k)	Delta vs Llama2 Cin + LibGen-v1 (step 48k)	Delta vs Llama2 Cin + LlbGen-v1 (step 51k)
hellaswag.0_shot.acc_char	70.35	67.64	67.95	2.72	2.40
math.4_shot.1_gen.em	1.76				
nq.5_shot.em	18.25	15.32	15.26	2.94	2.99
tqa.5_shot.em	45.50	42.35	40.61	3.15	4,89
piqa.0_shot.acc_char	76.82	74.92	76.50	1.90	0.33
siqa.0_shot.acc_char	47.34	47.19	46.93	0.15	0.41
mmlu.5_shot.macro_avg.acc_char	25.07	25.94	27.36	-0.87	-2.29
human_eval.0_shot.1_gen.em	2.44	2.44	2.44	0.00	0.00
arc_challenge.0_shot.acc_char	40.52	37.68	38.71	2,83	1,80
ppl.code_py	4.02	4,42	4,40	0.40	0.38

Locations:

- 7B Llama2 + LibGen v2 + OWM:

A: tair_ii

- 7B Llama2 Dill:

fair_llm/xldumps/az-230913_211008-gpt4tok/az-230913_211008-gpt4tok_run000/eval/0042500

- 7B Llama2 Cin + Libgen-v1

hikbash/eval_results/torchx-pci_7b_tok_cl100k_512_4m_with_libgen_v

14.09.2023

Jacob Xu run minhash deduplication of scitech, fiction and scimag:

NAME OF THE PARTY OF	P. C.	Part of the same of	I far the same and the same and the	Total Control Table Control
LibGen Part	Clean tokens	Minhash deduped	% duplicates removed	Location deduped
LIDGER Part	Clean tokens	wiiriiasii deduped	76 dupilcates removed	Location deduped

Sci-tech EN	220B	190B	15%	data/minhashdeduped/lib/sci tech/
Fiction EN	110B	70B	35%	data/minhashdeduped/lib/fic
Sci-mag EN	325B	320B	5%	data/minhashdeduped/lib/sci
Overall	655	560B	15%	

13.09.2023

Run ablation experiments for Sci-mag. Wandb: https://fairwandb.org/fairllm

Targeting 10T datamix with 3258 tokens from Sci-mag will make the 1x share of Sci-mag (LibGen papers) to be ~3.3%. To get more signal, we'll assume 2x epochs share in the final datamix, i.e. ~6.5% share.

So the ablation experiment would be to have the Dill datamix + LibGen papers 6.5% (with reducing proportionally CC share): config.

```
python stool.py relaunch
dumps/nb_78_libgen_papers_230913/nb_78_libgen_papers
_230913_ --exclude rsclearn[2662] --
launch_restart_dependencies 4
```

26.07.2023

Moved processed sci-mag to S3:

fair-use/scimage

20.07.2023

```
Re-run LibGen against a new 2k context length Dill baseline datamix:
                                                                                                                                                                                           Commented [2]: Where are we logging results for
                                                                                                                                                                                           this? any more details on the experiment?
python stool.py run
                                libgen_230720 train.py -- sweep
                                                                                           data_ablations/230616-fair-use-
                                                         --mem 480 --ncpu 10 --ngpu 8 --ntasks 256 --ncdes 32 --partition learn --anaconda
                                                                                                                                                                                           Commented [3]: the main results are below (04.07.2023), this was for the new baseline, but we
lib/230720 fair use lib en 7B b4M 256gpu.
               Fair_llm/
                                                     -- qos fair 11m -- launch restart dependencies 2
                                                                                                                                                                                           recently changed it to 4k context length, so this run is
                                                                                                                                                                                           not relevant (and was stopped).
 python stool.py run
                                 libgen 230720 train.py -- sweep
                      data_ablations/230616-fair-use-
                                                                                                                                                                                           I will schedule a new run on the new 4k Dill baseline.
                                                                                                                                                                                           But we can also use the previous runs (04.07.2023) - they showed positive signals.
 lib/230720 fair use lib en 7B b4M 256gpu.
                                                          --mem 480 --ncpu 10 --ngpu 8
  --ntasks 256 --nodes 32 --partition learn --anaconda
                                                                                                                                                                                           Commented [4]:
                           envs/xlformers_230705 -- gos fair_llm --
                                                                                                                                                                                           Commented [5]:
 launch_restart_dependencies 2
```

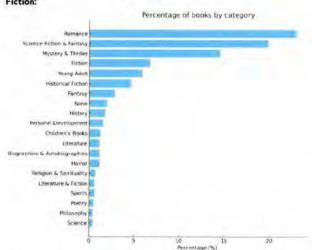
10.07.2023

Re-running evals for mmlu

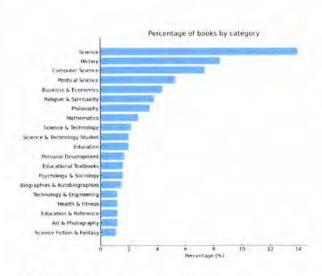


Categorisation of the data (performed by chatLLaMA):

Fiction:



Scitech:



04.07.2023

Ablation experiments results:

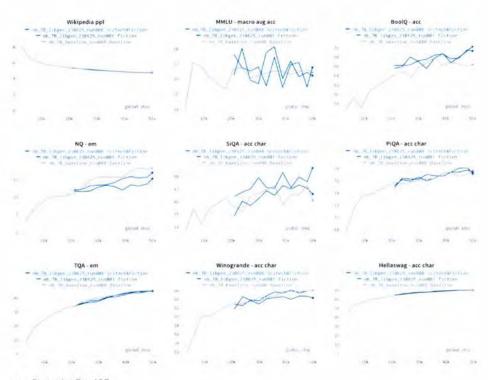
For experiments Exp 1 (Scitech+Fiction) and Exp 2 (Fiction only) we've substituted part of CCNET with LibGen to see the relative impact of the library to the baseline datamix. We observe improvements in the number of metrics:

- +4.5% BoolQ (+6% for Exp Fiction only)
- . +5.5% SiQA (+1.1% for Exp Fiction only)
- +1.2% MMLU

Next steps:

- Running Exp4: substituting both C4&CCNET with 2 epochs of LibGen. Hypothesis is that we can increase the number of epochs for LibGen.
- Running Exp5: substituting both C4&CCNET with LibGen in similar proportions. This would be a baseline for Exp4





Restarting Exp 4&5



01.07.2023

Ablation Experiments for LibGen:

Exp 1: Libgen Scitech + Fiction
 Sweep: https://github.com/fairinternal/
 Dir exp
 Fair_llm/xldumps/nb_7B_libgen_230625/nb_7B_libgen_230625_run000
 Exp 2: Libgen Fiction only
 Sweep: https://github.com/fairinternal/
 Dir exp:
 Fair_llm/xldumps
 Exp 3: Libgen vs B3G&Arxiv
 Sweep: https://github.com/fairinternal/
 Dir exp:
 fair_llm/xldumps/
 Exp 4: Libgen x2 epochs Scitech+Fiction

- Sweep: 230704_fair_use_lib_en_7B_b4M_256gpu.
- Direxp:/checkpoint/fair_llm/
- Exp 5: Libgen vs C4&CCNET
 - o Sweep: 230704_fair_use_lib_en_7B_b4M_256gpu.yaml
 - o Direxp: fair_llm/xldumps/

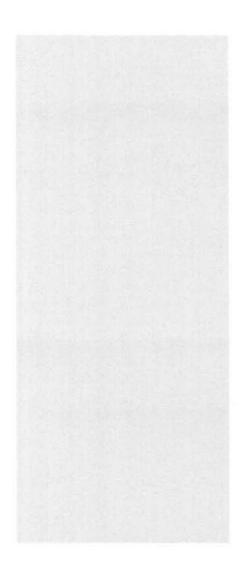
Exp 3&5: run command

python stool.py run nb_7B_libgen_230704 train.py --sweep 230616-fair-use-lib/230704_fair_use_lib_en_7B_b4M_256gpu. --mem 480 --ncpu 10 --ngpu 8 --ntasks 256 --nodes 32 --partition learn --anaconda fair_llm_pretrain --launch_restart_dependencies 2

30.06.2023

Statistics on OCR parsing failures:

AVG/doc	fiction_pdf	scitech_pdf	scimag_pdf
num_pages_per_book	170	258	6
num_chars_per_book	344,488	697,960	27,793
num_missing_page_fail_per_book	1.67 page / doc	11.2 page / doc	0.68 page / doc
num_missing_page_post_per_book	0.42 page / doc	14 page / doc	0.05 page / doc
errors_per_char	1.63E-05	7.23E-05	4.21E-05



- · Added parsed scitech_pdf and fiction_pdf with markers to determine the page break;
 - Fiction: air_llm/data/shuffled/libgen/fiction.
 - Scitech: Scitech:
 - Marker: "[MISSING_PAGE_*]":
 - MISSING_PAGE_EMPTY
 - MISSING_PAGE_FAIL
 - MISSING_PAGE_POST

MISSING_PAGE_EMPTY: (or almost empty) pages. In that case the model tends to collapse into a repetition very quickly. We are catching them at runtime but not always because communication is difficult there. The ones that get through will be caught by the POST processing in the very most cases

MISSING_PAGE_FAIL: the model will fail unexplainably somewhere in the page and diverge into a loop. It's determined by a heuristic with a constant threshold so there will be some that will be missed by that. These ones are then caught in the POST processing again.

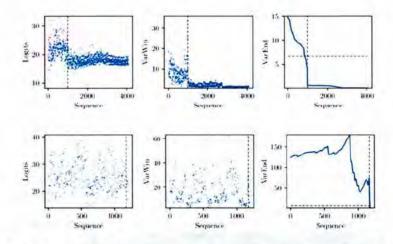


Figure 6: Examples for repetition detection on logits. Top: Sample with repetition, Bottom: Sample without repetition. Left: Highest logit score for each token in the sequence $\ell(x)$. Center: Sliding window variance of the logits $\mathrm{VarWin}_B[\ell](x)$, Right: Variance of variance from the position to the end $\mathrm{Var}\,\mathrm{End}_B[\ell](x)$

28.06.2023

[Nikolay]

- Relaunching failed ablation jobs (failed b/c of a bug in the xlformers):
 - fair_llm/xldumps/nb_7B_libgen_230625/nb_7B
- fair_llm/xldumps/nb_7B_libgen_230625/nb_7B
- W&B dashboard: https://fairwandb.org/fairllm/



19.06.2023

Starting an ablation experiment for 100% of EN scitech/fiction (330B tokens). We substitute 10% from CCNet with Libgen scitech dataset (matching it to the target datasets proportion: 2.3T Total vs 330B fiction/scitech -> scitech/fiction is 15%).

Experiments:

- Exp 1: EN: scitech+fiction
 - total dataset: 2.3T tokens
 - scitech&fiction is 330B tokens -> 15%
- Exp 2: EN: fiction
 - total dataset: 2.1T tokens
 fiction is 110B tokens -> 5%



Data:

Data	Total dataset	Baseline	Exp 1	Exp 2	Exp 3	Exp 4	Exp 5	Exp 6	Epochs
	size (billion	(weights/	(weights	(weights	(weights	(weights	(weights	(weights	(# / 200B)
	tokens)	%)	/%)	/%)	1%)	/%)	/%)	1%)	

Stack Exchange	25	1.2 (1.8%)						2.2	0.14
B3G (books3 + gutenberg)	28	3 (4.5%)			0			3.6	0.3
Arxiv	33	1.6 (2.4%)			0			2.8	0.15
Github OSS	271	3 (4.5%)						11.6	0.03
C4 en	198	10 (15%)			× -	6 (9%)	7 (10%)	7.7	0.15
CCNet	1,416	45 (67%)	35 (52%)	41.6 (62%)	39.6	29 (43%)	38 (57%)	27.4+ 32.8	E1: 0.07 E2: 0.09
Wikipedia	33	3 (4.5%)						4.3	0.27
Exp 1: Libgen Scitech + Fiction (nb_7B_libgen_2 30625_run000)	330B		10 (15%) sci: 6.6 fic: 3.4						0.09
Exp 2: Libgen Fiction only (nb_7B_libgen_2	110B			3.4 (5%)					0.09

30625_run001)							
Exp 3: Libgen vs B3G&Arxiv	330B		10 (15%) sci: 6.6 fic: 3.4				0.09
Exp 4: Libgen x2 Scitech+Fiction (nb_7B_libgen_2 30704_run000)	30B			20 (30%) sci: 13.2 fic:6.8			2
Exp 5: Libgen vs C4&CCNET (nb_7B_libgen_2 30704_run001)	30B				10(15%) sci: 6.6 fic: 3.4		1
Exp 6: Libgen - scimag (nb_7B_libgen_p apers_230913_r un000)						scimag: 6.5	
Total	Exp 1,3,5: 2.3T Exp 2: 2.1T Exp 4: 2.7T	67					

Run command (Exp1&2):

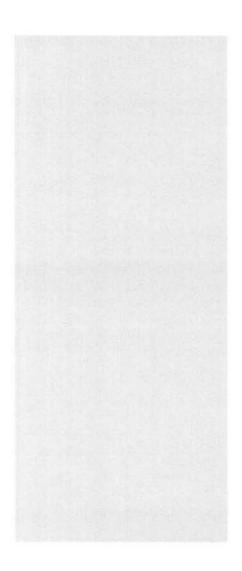
python stool.py run nb_7B_libgen_230625 train.py -- sweep

12-16.06.2023

[Nikolay]

- Total conversion (download -> cleaned):
 - Scitech: 82% (b/c most of scitech are PDFs)
 - Fiction: 86%
 - Scimag: TBD
- non-EN languages:

Language (Sci- tech)	Share, % (Sci-tech)	Language (Fiction)	Share, % (Fiction)
Spanish	23.4%	French	23.1%
Italian	16.0%	German	22.7%
Chinese	13.3%	Spanish	15.0%
Portuguese	11.9%	Dutch	10.5%



German	10.5%	Italian	8.2%
French	7.4%	Hungarian	5.0%
Russian	3.7%	Portuguese	3.5%
Hungarian	2.2%	Chinese	2.8%
Dutch	1.7%	Japanese	2.2%
Turkish	1.1%	Czech	1.5%
Other	8.8%	Other	5.4%

07.06.2023

[Nikolay]

- Added script to convert .mobi to .epub to further parse with epub2markdown script (~60k additional documents, ~10B tokens).
- Converted 7k scitech .mobi to .epub (5% of scitech non-en)

06.06.2023

[Nikolay]

Done with the EN Scitech/Fiction part. Now finishing the non-EN Scitech/Fiction and ALL Scimag.

- Sci-tech (non-en):
 - Downloaded 130k (99%) of non-English epub/mobi Sci-tech books and 586k non-English epub/mobi Fiction books

- We decided to skip the PDFs for now (since it'll be a hard lift to parse them with our current OCR). There are ~1M non-En PDFs, 60% of which are in Russian (which is not our target language), so the remaining is 425k PDF books (~65B additional multi-lang tokens) which we skip.
- Fiction (non-en):

Libgen Part (non-EN)	Total non-EN (num)	Downloaded (num / %)	Parsed (num / %)	Location Raw	Location Parsed
Sci-tech EPUBs	130,593 (123,281 epub + 7,312 mobi)	128,722 / 99%	0 / 0%	fair- use/scitech/epub_non_en/	air_llm/data _v2/datasets/books
Fiction EPUBs	594,348 (545,578 epub + 48,770 mobi)	586,240 / 99%	0 / 0%	fair- use/παιοη/ερμο_non_en	air_llm/data _v2/datasets/books/
Sci-mag All (incl EN)	81,903,411 (876 chunks)	690 / 79%	100 / 11%	/fair- use/scimag/	

05.06.2023

[Nikolay]:

- Sci-mag is 70% downloaded
- Downloaded the remaining 5% of Sci-tech, but all corrupted (unable to parse)
- Parsing multi-lang scitech/fiction PDFs seems to be quite time-consuming we need to re-train OCR parsing script (no-immediate training data for that), so we'll start with EPUB/MOBI formats for non-English books
- Started loading multi-lang Scitech & Fiction:
 - Fiction (non-en, non-pdf): epub=545,578, mobi=48,770

- Scitech (non-en, non-pdf): epub=123,281, mobi=7,312
- Convert Scitech multi-lang EPUBs to markdown to check the quality of conversion (could be used for training the OCR for multi-eng)

[Lukas]:

- We can get additional 8-9% of non-English Sci-tech PDFs (~400k books). But for that we need training data for Spanish, German, Italian, French (optional: Chinese, Portuguese):
 - Check if we have training data on Arxiv

Libgen Part (EN)	Total EN (num)	Downloaded (num / %)	Parsed (num / %)	Location Raw	Location OCR Parsed	Location Cleaned	Cleaned tokens (#)
Sci-tech PDFs	1,272,655	1,241,150 / 98%	1,104,047 / 89%	fair- use/scitech/pdf_en/	/large_experiments/fair_llm	/fair- use/scitech/processed	1708
					fair- use/scitech/pdf_ocr/		
Sci-tech EPUBs	454,064	454,534 / 100%	392,426 / 86%	fair- use/scitech/epub_en/	fair_llm/da ta_v2/datasets/	air-use/ scitech/processed/202	50B
Fiction PDFs	117,980	106,362/ 90%	96,981 / 82%	fair- use/fiction/pdf_en/	fair_llm/da ta_v2/datasets/books,	fair- use/fiction/processed/	10B
					use/fiction/pdf_ocr/		
Fiction EPUBs	1,041,740	1,031,934 / 99%	946,144 / 91%	fair- use/fiction/epub_en/	fair_llm/da ta_v2/	fair-use/ fiction/processed	100B

					ion_epub_en_processed_	
Sci-mag All	81,903,411 (876 chunks)	619 / 70%	100/11%	s3://fairspark-data/fair- use/scimag/	fair- use/scimag/pui_oct-/	

01.06.2023

Libgen Part (EN)	Total EN (num)	Downloaded (num / %)	Parsed (num / %)	Location Raw	Location Parsed	Location Cleaned	Cleaned tokens (#)
Sci-tech PDFs	1,272,655	1,165,867 / 92%	1,103,695 / 95%	fair- use/scitech/pdf_en/	fair_llm/da ta_v2/datasets/books/	use/scitech/processed/20230 526_pdf_en/	170B
Sci-tech EPUBs	454,064	454,534 / 100%	392,426 / 86%	fair- use/scitech/epub_en/	fair_Ilm/da ta_v2/datasets/books	air-use/ scitech/processed/20230526_ epub_en/	50B
Fiction PDFs	117,980	105,077/ 89%	96,981 / 82%	/fair- use/fiction/pdf_en/	fair_llm/da ta_v2/datasets/books/data/fict ion_pdf_ocr_all	fair- use/fiction/processed/202305 26_pdf_en/	10B
Fiction EPUBs	1,041,740	1,022,914 / 98%	946.144 / 91%	fair- use/fiction/epub_en/	fair_llm/da ta_vz/datasets/books.	fair-use/ fiction/processed/20230526_e pub_en/	100B
Sci-mag All	81,903,411 (876 chunks)	451 / 51%	24 / 3%	fair- use/scimag			

[Lukas]

- Started with Scimag
- Optimized Nougat OCR inference code for many small documents

30.05.2023

Libgen Part (EN)	Total EN (num)	Downloaded (num / %)	Parsed (num / %)	Location Raw	Location Parsed	Location Cleaned	Cleaned tokens (#)
Sci-tech PDFs	1,272,655	1,165,867 / 92%	1,066,478 / 91%	fair- use/scitech/pdf_en/	fair. Ilm	fair- use/scitech/processed/20230 526_pdf_en/	170B
Sci-tech EPUBs	454,064	454,534 / 100%	392,426 / 86%	fair- use/scitech/epub_en/	air_IIm/da ta_v2/datasets/books/data,	fair-use/ scitech/processed/20230526_ epub_en/	50B
Fiction PDFs	117,980	105,077/ 89%	96,981 / 82%	fair- use/fiction/pdf_en/	air_llm/da	fair- use/fiction/processed/202305 26_pdf_en/	108
Fiction EPUBs	1,041,740	1,022,914 / 98%	946,144 / 91%	fair- use/fiction/epub_en/	fair_Ilm/da ta_v2/datasets/books	fair-use/ nction/processed/20230526_e pub_en/	100B
Sci-mag All	81,903,411 (876 chunks)	435 / 50%	0	fair- use/scimag			

Libgen Part (EN)	Total EN (num)	Downloaded (num / %)	Parsed (num / %)	Location Raw	Location Parsed	Location Cleaned	Cleaned tokens (#)
Sci-tech PDFs	1,272,655	1,165,867 / 92%	1,066,478/ 91%	fair- use/scitech/pdf_en/	fair_llm/da ta_vz/gatasets/books/data,	fair- use/scitech/processed/20230 526_pdf_en/	170B
Sci-tech EPUBs	454,064	454,534 / 100%	392,426 / 86%	/fair- use/scitech/epub_en/	fair_Ilm/da ta_v2/datasets/books/data	fair-use/ scitech/processed/20230526_ epub_en/	50B
Fiction PDFs	117,980	105,077/ 89%	96,981 / 82%	/fair- use/fiction/pdf_en/	fair_llm/da ta_v2/datasets/books/data	/fair- use/fiction/processed/202305 26_pdf_en/	10B
Fiction EPUBs	1,041,740	1,022,914 / 98%	946.144 / 91%	fair- use/fiction/epub_en/	fair_llm/da ta_v2/datasets/books/data	fair-use/ liction/processed/20230526_e pub_en/	100B
Sci-mag All	81,903,411 (876 chunks)	361 / 41%	0	fair- use/scimag			

[Peter]

- Also had memory limitations
- Finalized book filters:

Condition	Example of an affected book
Book line count less than 50	# Table of Contents
	1. Cover
	2. Title Page
	3. You Can Be Brave

	## Guide			
	1. Start Content			
	# Table of Contents			
	1. Cover			
	2. Title Page			
	3. You Can Be Brave			
	## Guide			
	1. Start Content			
Non-empty lines have less	# Guide			
than 20 characters avg	1. Cover			
length				
	2. Text			
	# Page Numbers			
	1.1			
	2.2			
	3.3			
	4.4			
	5. 5			
	6.6			
	7.7			
	8. 8			
	9.9			
	19. 19			
	20. 20			



HIGHLY CONFIDENTIAL - SOURCE CODE

Meta_Kadrey_00065281

	21. 21 22. 22 23. 23 24. 24 25. 25 26. 26
Numeric fraction of characters > 10%	1. 2 2. 3 3. 4 4. 5 5. 6 6. 7 16. 17 17. 18 18. 19 19. 20 20. 21
Line longer than 50k characters	Book without any new lines or formatting, sometimes a parsing issue
Language id less than 0.5 for english	Our pdf ocr model is trained on english documents, so there are hallucinations when ocring non-english text. Also we only want english book for now.

HIGHLY CONFIDENTIAL - SOURCE CODE

P.- J. HÉRAULT
CAL DE TER
COLLECTION
« ANTICIPATION »
ÉDITIONS FLEUVE NOIR
6, rue Garantière – PARIS VIe

Scimag:
fair-use/scimag

Stat for filtering Fiction_epub:

Total number of books processed: 945531

Metrics for the number of books filtered out:

- book_line_count: 4951 books (0.52% of total books)
- book_length: 1928 books (0.20% of total books)
- numeric_fraction: 261 books (0.03% of total books)
- -long_line: 3362 books (0.36% of total books)
- non_english: 5602 books (0.59% of total books)

Metrics for the average number of lines removed:

- repeated_lines: 0 lines per book on average
- missing_page_markers: 0 lines per book on average
- removed_boilerplate: 97 lines per book on average
- stripped_lines: 4 lines per book on average



Aggregate Metrics:

- Total number of books removed: 11249
- Percentage of books removed: 1.19%

Downloaded: 1,022,914

After parsing errors and filtering: 946,144 (~5% lost due to not being able to parse epubs, 1% through filtering)

Scitech_pdf_ocr_all:

Total number of books processed: 1060234

Metrics for the number of books filtered out:

- book_line_count: 12422 books (1.17% of total books)
- book_length: 5684 books (0.54% of total books)
- numeric_fraction: 5695 books (0.54% of total books)
- long_line: 70 books (0.01% of total books)
- non_english: 17902 books (1.69% of total books)

Metrics for the average number of lines removed:

- repeated_lines: 0 lines per book on average
- missing_page_markers: 37 lines per book on average
- removed_boilerplate: 65 lines per book on average
- stripped_lines: 1 lines per book on average

Aggregate Metrics:

- Total number of books removed: 27677
- Percentage of books removed: 2.61%

HIGHLY CONFIDENTIAL - SOURCE CODE

Meta_Kadrey_00065284

[Nikolay] Had memory limitation on fair cluster of (20T) so had to back up everything to s3:

- Fiction
 Fair-use/fiction
- Scitech: /fair-use/scitech
 Scimag: /fair-use/scimag
- 24.05.2023

[Lukas]

- Script to filter SciMag files (script):
 - · Checks if file is corrupt
 - · Checks if file is PDF
 - . Checks if PDF text is english
 - → Send to Nougat OCR

Libgen Part (EN)	Total EN (num)	Downloaded (num / %)	Parsed (num / %)	Location Raw	Location Parsed	Location Cleaned
Sci-tech PDFs	1,272,655	1,072,286 / 84%	1,025,070 / 81%	fair- use/scitech/en_pdf	fair_llm/	
Sci-tech EPUBs	454,064	454,534 / 100%	392,426 / 86%	fair- use/scitech/en_epub		
Fiction PDFs	117,980	102,118/ 87%	96,981 / 82%	/fair- use/fiction/en_pdf	air_llm/	

Fiction EPUBs	1,041,740	1,001,538 / 96%	642,703 / 64%	fair- use/fiction/en_epub	
Sci-mag All	81,903,411	37,713 / 0%	0	use/scimag	

Notes:

- Trying to load scimag with the same approach (direct download) as before - doesn't seem to be fast (250k docs / 12h -> 160 days to download the library). Exploring other options to load faster.

			n 14 (n)	700 S00000 2000	versex agency
Libgen Part (EN)	Total EN (num)	Downloaded (num / %)	Parsed (num / %)	Location Raw	Location Parsed
Sci-tech PDFs	1,272,655	1,179,045 / 93%	973,340 / 76%	fair-use/scitech/en_pdf	fair_llm/data_v2/datasets/books
Sci-tech EPUBs	454,064	454,534 / 100%	392,426 / 86%	fair-use/scitech/en_epub	scitech_
Fiction PDFs	117,980	102,118/ 87%	69,423 / 59%	fair-use/fiction/en_pdf	fair_IIm/data_v2/datasets/books
Fiction EPUBs	1,041,740	1,001,538 / 96%	642,703 / 64%	fair-use/fiction/en_epub	data/fiction
Sci-mag All	81,903,411	0	0	fair-use/scimag	

Notes:

- On the weekend hit the hard limit of disk utilization on fair cluster: ~24T (in my personal folder nikbash)
- Had to clean the disk (what was possible to clean), so now around ~21T
- With these constraints can't easily load scimag (~80T), so
 - EITHER distribute download across team (we have same IP, so would be throttled by libgen)
 - OR transfer raw files to S3, remove them from fair cluster (need to finish processing first) and load scimag in chunks
 - OR increase the disk space
- The problem with scimag loading is that there is no metadata for it, so we can't pre-filter by language and extension first, so we need to load everything at once (in chunks)
- Started backing up raw data to S3 bucket (to further remove raw data from the fair cluster)

Fiction:	
- EPUBs:	fair-use/fiction/epub_en/
- PDFs:	fair-use/fiction/pdf_en/
Scitech:	
- EPUBs:	fair-use/scitech/epub_en/
- PDFs:	fair-use/scitech/pdf_en/

Libgen Part (EN)	Total EN (num)	Downloaded (num / %)	Parsed (num / %)	Location Raw	Location Parsed
Sci-tech PDFs	1,272,655	1,179,045 / 93%	904,149 / 71%	fair-use/scitech/en_pdf	fair_llm/data_v2/datasets/books
Sci-tech EPUBs	454,064	454,534 / 100%	392,426 / 86%	air-use/scitech/en_epub	data/scitech_
Fiction PDFs	117,980	102,118/ 87%	67,192 / 57%	fair-use/fiction/en_pdf	fair_Ilm/data_v2/datasets/books

Fiction EPUBs	1,041,740	1,001,538 / 96%	642,703 / 64%	/fair-use/fiction/en_epub	data/fiction
Sci-mag All	81,903,411	0	0	air-use/scimag	

Notes:

- . The download speed dropped significantly for the remaining 15% of data (probably the data is on the servers with low throughput)
- · Planning to start loading Sci-mag on the weekend
- . Discussed with Lukas Blecher that we would need to train the Nougat OCR on other languages to be able to parse the non-EN PDFs somewhere around end of June, 23
- Started parsing Fiction PDFs with Nougat OCR:
 - The quality of other PDF parsers was not satisfactory (see notes from 18.05.2023)
 - The number of EN PDFs in Fiction is relatively small 117k, so we need just 1-2 days with 500 GPUs

Libgen Part (EN)	Total EN (num)	Downloaded (num / %)	Parsed (num / %)	Location Raw	Location Parsed
Sci-tech PDFs	1,272,655	1,006,428 / 80%	822,167 / 65%	fair-use/scitech/en_pdf	air_llm/data_v2/datasets/books
Sci-tech EPUBs	454,064	454,356 / 100%	392,426 / 86%	fair-use/scitech/en_epub	data/scitech_
Fiction PDFs	117,980	69,554 / 59%	0	fair-use/fiction/en_pdf	
Fiction EPUBs	1,041,740	780,513 / 78%	642,703 / 64%	fair-use/fiction/en_epub	data/fiction_
Sci-mag All	81,903,411	0	0	fair-use/scimag	

Libgen Part (EN)	Total EN (num)	Downloaded (num / %)	Parsed (num / %)	Location Raw	Location Parsed
Sci-tech PDFs	1,272,655	1,006,428 / 80%	645,061 / 51%	fair-use/scitech/en_pdf	fair_llm/data_v2/datasets/books
Sci-tech EPUBs	454.064	454,292 / 100%	392,426 / 86%	fair-use/scitech/en_epub	flibgen_epub_
Fiction PDFs	117,980	67,274 /	0	/fair-use/fiction/en_pdf	

		50%			
Fiction EPUBs	1,041,740	696,542 / 70%	642,703 / 64%	/fair-use/fiction/en_epub	data/fiction
Sci-mag All	81,903,411	0	0	/fair-use/scimag	

Notes:

- We looked at processing the Fiction PDFs with a non-ocr parser PYPDF2, as it would be much faster. But even with normal novels there are lots of artifact like missing spaces or random spaces within words
- Therefore we decided to also use nougat our for all the fiction pdfs

PYDF2 (with spacing issues)	Nougat OCR

Chapter

It'sbeen inmypocket theentire time. Lending meacomfort theorigins ofwhich Ihad temporarily forgotten. Iremember itnow, and slowly. I begin to realize I might live.
Pulling itoutofmypocket, Iseehow itreflects thestrange, dim, purple light officeoffin like room I've been confined to.Ialmost put myself into atrance looking atti, and playing short films inmy head ofhow Immy employ it. The walls offins room arccurved and feel like skin. Icanfeel avibration thrumming throughout, like adistant, powerful, engine. I'emotsure how long I've been mying here, I'm notewer sure how long I've been make. Itseens I just realized over time Iwas conscious and thinking. After what feels like 20minutes of just staring atthe reflected purple light Texplore the walls offits room, looking foranopening, shandle, apuff ofirtelling mel'mot

For Luthor, my purpose.Chapter 1 It's been in my packet the entire time. Lending me a confort the origins of which I had temporarily forgotte
**Pulling it out of my pocket, I see how it reflects the strange, dim, purple light of the coffin like room I've been confined to. I almost put m
Mo dir, no tock.

**The purple light has no source that I can lind; it seems to evenly enamate from the fleshy walls of my prison. I puth on these walls, and find
**I try to calm myself before I coemit the act that's probably going to lead to me getting killed, or at the very least, tortured again. In makin
**Looking back at my hunds, and once more at the item still umbelievably with me, I don't feel the least bit absure asking for strength and coura
Fine to begin Step One * **Lying on my side, using my left hund, I drive the blade of my screwdriver into the wall of the flesh-like substanc
I need to be more careful.

Chapte r One: First N ight

The dungeon door slam med shut be hind her. His eyes glow ed yel "So you're what they've found for me. Yo u can come closer. I'm He was propped up on pillows at the head of a large four- poste room. Her eyes adjuste d to the near-darkne ss. She could just beside the bed, on it s ome roast ed meat... fruit...wine...and be fo the bed dominated the room, so t he man dominated the bed. He w closer. Man acles tightly wrapped his wrists and were attached t to the upper be dposts . Similar c hains on t he foot posts dis indicating his feet w ere als o chained to the bed. T he firelig highlightin g a face of predatory male beauty: high cheekbones, straight nos e above a beautifully shaped mouth. His long hair shoulders to mid chest. Nake d, dark hone y skin covered his w abdomen. S he had the oddest urge to pull back the cove r and s her hand to control the impulse.

His gaze ret urned her frank assessment. She knew he would see a share d ancestry. Her da rk hair was pulled back in a loose bra

Chapter One: First Night

The dungeon door slammed shut behind her. His eyes glowed yellow in the firelight

"So you're what they've found for me. You can come closer. I'm bound...for now."

"So you're what they've found for me. Yo u can come closer. I'm
He was propped up on pillows at the head of a large four-poste
room. Her eyes adjusted to the near-darkness. She could just
make out a small table beside the bed, on it some roasted meat...Inuit...wine...and before the tire a small rug. As the bed dominated the room, so the man
dominated the bed. He was huge. She dared a step closer. Manacles tightly wrapped his wrists and were attached to chains that bound his arms to the upper
beside the bed, on it some roast ed meat... fruit...wine...and be fo
the bed dominated the room, so the man dominated the bed. He was
huge. She dared a step closer. Manacles tightly wrapped his wrists and were attached to chains that bound his arms to the upper
beside the bed, on it some roast ed meat... fruit...wine...and be fo
highlighting a face of predatory male beauty; high cheekbones, slightly tilted eyes and a long straight nose above a beautifully shaped mouth. His long hair
appeared black and trailed over shoulders to mid chost. Naked, dark honey skin covered his well-muscled chost and abdomen. She had the oddest urge to pull
back the cover and see what tay beneath and fisted her hand to control the impulse.

As of 5pm

Libgen Part	Total EN (num)	Downloaded (num / %)	Parsed (num / %)	Location Raw	Location Parsed
Sci-tech PDFs	1,272,655	835,499 / 65%	645,061 / 51%	libgen_pdf	air_IIm/data_v2/datasets/books
Sci-tech EPUBs	454,064	454,292 / 100%	0	libgen_epub	data/libgen
Fiction PDFs	117,980	58,071 / 49%	0	fiction/fiction_pd	f
Fiction EPUBs	1,041,740	627,218 / 60%	0	fiction/fiction_ep	ub
Sci-mag All	81,903,411	0	0		

16.05.2023

[Lukas]

Sci-Tech conversion status (6pm 16.05.2023): (38% done of 1,726,719)

PDFs (579,620 or 46% of 1,272,655):

fair_llm/data_v2/datasets/books/

• EPUBs (82,699 or 18% of 454,064):

data/libgen

[Nikolay]

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Scitech EN download status (6pm 16.05.2023): (95% done of 1,201,994)

Fiction EN download status (6pm 16.05.2023): (55% done of 1,159,720)

EPUBs (580,899)
 iction/fiction_epub

PDFs (55,633):

Robert Stojnic suggested that we could do an experiment with finetuning 70B model on the sci-tech data to check that it would improve the reasoning capabilities (ideally to match the Galactica):

Option	GPU hours	Comment
70B on 512 GPUs	362h (15 days)	sci-tech tokens (1 epoch): 200B wps 708: 300
70B on 1024 GPUs	181h (7.5 days)	GPU*hours = 200B/(num_g*wps*3600s)
70B on 2048 GPUs	90h (3.7 days)	

15.05.2023

[Lukas]

Sci-Tech conversion status (5pm 15.05.2023): (34% done of 1,726,719)

PDFs (499,404 or 39% of 1,272,655):

EPUBs (82,699 or 18% of 454,064):

data/libgen_epub_parsed

SciMag calculation:

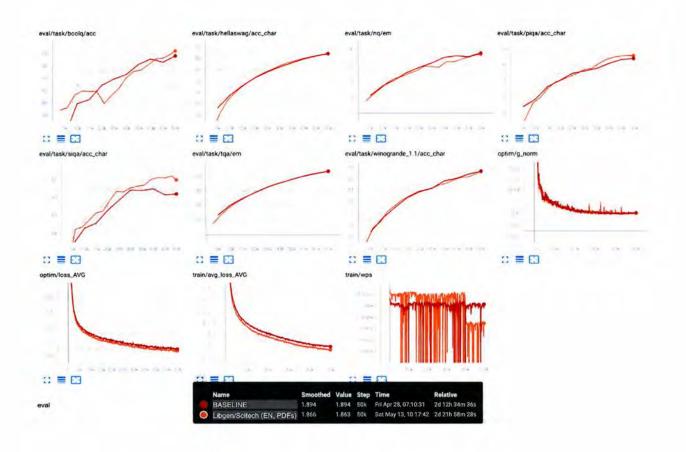
Processing speed: (12.6±10.5) s/batch @ 4 pages per batch #pages SciMag: 50%*82M*6=246M pages (assume 50% english)

Estimated GPU hours: (12.6±10.5)*246M/4/3600= (215±180)k GPUh [Nikolay] Instructions to download libgen: fair_data/fair_data/projects/fair_use_lib · in your fair cluster terminal run "screen -S fiction" · in a new screen window; o source activate _libgen_direct.py" Scitech EN download status (12pm 15.05.2023): (95% done of 1,201,994) • EPUBs (310k): 111,292 on FAIR Cluster: bgen_epub 199,145 on l PDFs (847k): 647,932 on Fair Cluster: libgen_pdf o 199,145 on loaded previously EN PDF/EPUB on fair cluster (~480k); air_llm/data_v2/datasets/books Fiction EN download status (5pm 15.05.2023): (33% done of 1,159,720) EPUBs (338,797): fiction/fiction_epub fiction/fiction_pdf PDFs (44.109): Ablation results for Scitech EN PDFs at 50k step (100% complete): · Overall no red flags observed Some improvement on siga and boolg (but that's within the stdev)

. TB; https://fburl.com/

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Meta_Kadrey_00065294



HIGHLY CONFIDENTIAL - SOURCE CODE

Meta_Kadrey_00065295

[Nikolay]

We have overall downloaded 1,6M books EN PDFs and EPUBs for Scitech (or 92%). This number however contains ~10% of corrupted file which needs to be re-downloaded later on (or skipped if they are corrupted in the source)

Scitech EN download status (12pm 12.05.2023): 92% done

- EPUBs (305k)
 - 111,272 on FAIR Cluster
 - = 199,145 on RSC
- . PDFs (810k)
 - 638,350 on Fair Cluster
 - o 199,145 on RSC
- loaded previously EN PDF/EPUB on fair cluster (~480k).

fair_llm/data_v2/datasets/books

Ablation results at 35k step (70% complete):

- · Overall no red flags observed
- . Loss seems to flatten out earlier
- . TB: https://fburl.com

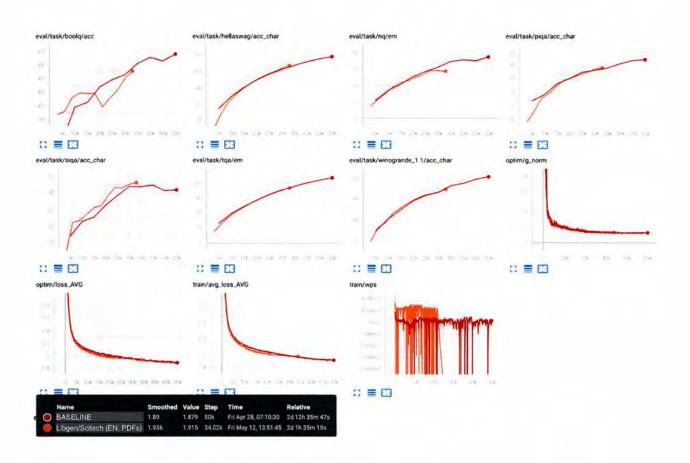
[Lukas]

Scitech EN PDF conversion status (2pm 12.05.2023):

- 350k books finished (52B tokens)
- · 390k books ready to process

[Nikolay] ablation results:

HIGHLY CONFIDENTIAL - SOURCE CODE Meta_Kadrey_00065296



HIGHLY CONFIDENTIAL - SOURCE CODE Meta_Kadrey_00065297

[Nikolay]

We will need to reload the corrupted files separately after going through the first round of parsing.

10% of files are corrupted files after initial download, both EPUBs and PDFs.

For EPUBs processing we used Marie-Anne's html2latex.py script (the one used for CC) and performed some post processing on top of it - removing the Copyright section.

Scitech download status (10pm 11.05.2023):

- EPUBs (305k)
 - o 106,677 on FAIR Cluster
 - o 199,145 on RSC
- PDFs (810k)
 - o 611,409 on Fair Cluster
 - o 199,145 on RSC
- loaded previously EN PDF/EPUB on fair cluster (~480k):

[Lukas]

PDFs:

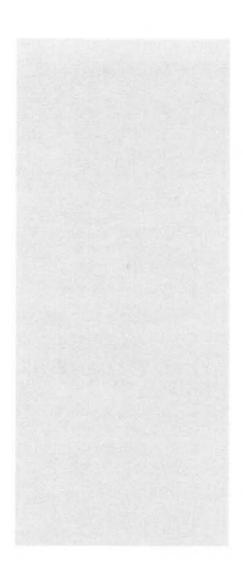
Improved post-processing to remove all kinds of repeated patterns and more.

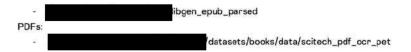
- 260k PDF books successfully parsed
- datasets/books/data/scitech_pdf_ocr_all

[Peter]

EPUBs: prepared a script to postprocess the EPUBs.

- 82k EPUB books parsed



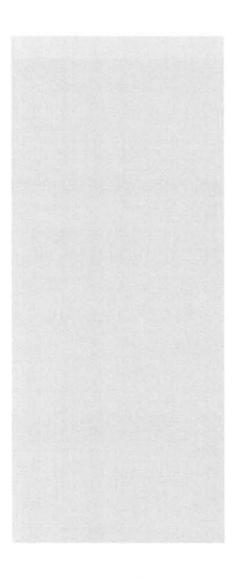


[Nikolay]

Libgen Scitech PDFs:

Starting an ablation experiment for 10% of scitech (parsed pdfs). We substitute 10% from CCNet with Libgen scitech dataset (matching it to the target datasets proportion: 2T Total vs 200B Libgen Scitech -> 10%).

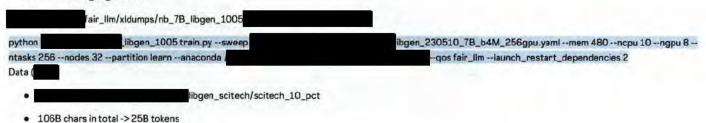
Data	Total dataset size (billion tokens)	Baseline (weights/%)	Experiment (weights/%)	Epochs (# / 200B)
Stack Exchange	25	1.2(1.8%)	1.2 (1.8%)	0.14
B3G (books3 + gutenberg)	28	3 (4.5%)	3 (4.5%)	0.3
Arxiv	33	1.6 (2.4%)	1.6 (2.4%)	0.15
Github OSS	271	3 (4.5%)	3 (4.5%)	0.03
C4 en	198	10(15%)	10 (15%)	0.15



CCNet	1,416	45 (67%)	38(57%)	0.08
Wikipedia	33	3 (4.5%)	3 (4.5%)	0.27
Libgen Scitech	25B (total: ~200B)		7 (10%)	0.8
Total	2.2T	67	100%	

Run (TB: https://fburl.com/

- Libgen 10%: nb_7B_libgen_1005_run000
- Baseline: nb_7B_baseline



Config: https://www.internalfb.com/

Libgen Scitech EPUBs:

The goal is to apply the same html_to_latex parser from CCNET.

html2text parsing (import html2text)	CC html_to_latex parsing (crawl.utils.html_to_latex)
PLAN B 3.0	PLANB30
Entering a New World During that late reminered if 2007, this means at an observating fice millifree amound as a threstoclosers, in many facilities reported. The Action core (an his collected at an intervendented rate into some a red base is a record found from the collected at an intervendented rate into some are of base of the an object of the collected at an intervendented rate into some are of base of base of base of base of the same at a record found into the base of the same of base of the same of a pain is against an intervented at a subject of base of the same of a pain is against an intervented at a subject of base of the same of the sa	seal of an investing to the day of the company of t

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Meta_Kadrey_00065301

Summary

Here are the key points we covered in this chicker:

- Facebook, the web and OS have time milts educate year besides yeing the most popular game platforms must at acceptance of the -budget game. Missolinus connectives social relative and permit in youther platforms.
- Fixenest by note that Name in menting zon, asset 30-40 percent propely gives on the continessed. This exident
 both gradient and all the major agreeming spirits.
- Add a 10th million person injured index of price with on an decar and other conspicus point in decirity. Only 20 procedular
 bases informations are more of the control on discussion on energy models. It is used in America, Romal, and Tarray
 2023 has a local form all located on more than 2021 of the configuration of the control of the control of person of the control of the contro

Charter 5

OS versus Facebook versus the Web: What's the Right Plutforn?

n This Chapter

- Reviewing what works and what doesn't entitle
- Reviewing what works and what doesn't an Pocebook
 Reviewing what works and what doesn't in web games.
- Registers the bodying traffer patters that an object of in pool here a return our bod. As they be the most distinct for the state of a most patter is required, particle that gour pro-designed transaction are out of them. Explicit additional points to use to be for a contributed about 10 per particle and history as all found that is dispersional points are out, a state is a bottom to the provision. The badde were all dispersional properties of making logical and going it wis to a commission of the contribution or provision of such forecordant. In other words was

permagnessly was better on or a platform than others, and the sea penglectual effort better opportunity for buscess. The official models of all the country are compared for the penalty of the others of a penalty of the country and the first of the country of the others.

is unan sinally burs their prociticants registered in the syntem, with many Floribook gamers already have a bank of virtual our rice

Reviewing What Works and What Doesn't on iOS

and which make them more lively in size of an your guine). At the same time, some guine

0C7955E04686D9FD4CF2C2FA5D1B390C

Summary

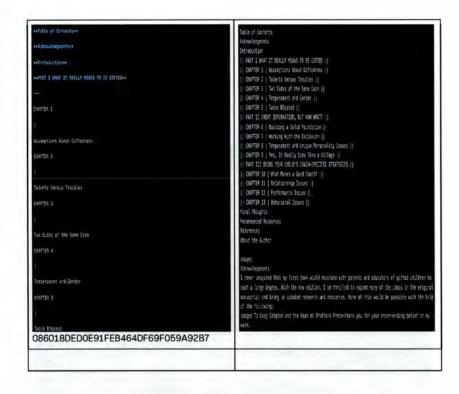
er and te king pår åve ste en dir dit sidigere. Fastansk for en ett, and Scherrichen esse aktorisjan heldes åvergit en est gold generalleringen stem kinden som ett at att gede goddigt et en er stem for ett att gede goddigt et en en ett gede goddigt ett en ett gede goddigt ett gede goddigt ett goddigt ett gede goddigt ett gede goddigt ett go

Ower 10 years before what is the Next of High Editoria Tool Couper - Booking with an extended of extended 10 - Booking hill works and work too from Paration - Booking with contributed to contribute by the Augusts in the late couper, and the second of the August of the book throw in assess, or the Augusts and Augusts in the second of the second of

Reviewing What Works and What Doesn't on iOS

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Meta_Kadrey_00065302



HIGHLY CONFIDENTIAL - SOURCE CODE Meta_Kadrey_00065303

[Nikolay]

Decided to go with the direct file upload without using torrents for the following reasons:

- . using torrents would entail "seeding" the files i.e. sharing the content outside, this could be legally not OK
- . with the direct file download we can pre-filter the needed format and language of the files i.e. downloading only EN, PDF and EPUB initially
- . the downside is that this way it is slower and need more engineering to bypass IP throttling and download retries
- we can reload specific MD5 file names, that were corrupted or missing from the initial download (based on Lukas's observations there are 30% of corrupted files in the initial Libgen download)

Currently loading using 2 dev machines and 1 fair cluster. Approximately an additional 10TB of data loaded (1M books out of 1.3M): 75% of EN, PDF or EPUB scitech books.

Raw downloaded data locations: • ~800k EN books on fair cluster: • 546k pdfs • 80k epubs • 8k corrupted files • ~400k EN books on • 200k pdfs • 200k epubs

loaded previously EN PDF/EPUB on fair cluster (~480k):

datasets/books

Parsed data:

10% scitech (pdfs only):

//ibgen/scitech_10_pct/

HIGHLY CONFIDENTIAL - SOURCE CODE

Total numbers (in # of books):

- Libgen: 3.7M
- Libgen (EN & PDF/EPUB): 1.7M | Downloaded 1.3M
- Libgen (EN & PDF): 1.3M -> parsed 13%

Examples of parsed EPUBs (light version of parsing w/o M-A's script):



[Lukas]

Filtered scitech conversion is 75% done out of the first chunk of 340k EN PDF books (total chunk size of scitech EN PDFs: 1.3M, so we've parsed ~13% of EN PDFs). We pre-selected 340k books (PDF). 34% of the files are corrupted. Finished 167k (uncorrupted) books.

Conversion speed:

- Ideal: 6.8 ± 1.9 PDF / GPU*h
- Actual (b/c of insufficient number of GPUs): 2.2 PDF / GPU*h

Commented [6]: That paints a wrong picture. The speed per GPU is still around 7 books per hour. The number of GPUs is the bottleneck

Commented [7]: fair point, can you give a ballpark how much more we need? I gmeta.com was mentioning that we can yet 1200 oPUs from the Retina team

Commented [8]: As a ball park estimate: we would need "2k GPUs on FAIR Cluster for 2 weeks starting from "mid-next week. It's actually similar to what we use now for sci-tech, so we might keep the current strategy of just asking people to help run from their accounts.

meta.com.

Implemented an additional step of post processing to remove repeated reference items.

Combined directory: scitech_pdf_ocr_all

Processed chunks (~10% of scitech):

books/data/scitech_pdf_ocr_jsonl/chunks

05.05.2023

Launched slurm jobs for OCR parsing of the first 15% of Libgen:

- scitech_pdf_ocr: first half of parsed files
- scitech_pdf_ocr_af: second half of parsed files

Slurm job command

04.05.2023

Plan:

- . [Nikolay] check about gpus on fair cluster -> how much we can use: 1k GPUs DONE
- . [Lukas] prepare 3 sample pages of books with formulas, tables and lists original VS parsed with small OCR model DONE
- . [Nikolay] Pre-filter data to only EN (since OCR parsing works best with EN) -> We will pre-filter EN and PDFs only as the OCR script works best with EN. DONE
- [Lukas][Nikolay] start the pipeline for parsing first 10% of PDFs on fair cluster: use the current dump of PDFs:
- . [Lukas] add the token to split the sequence (in case the page was skipped due to parsing error) DONE
- [Nikolay] prepare pipeline for loading remaining data from libgen DONE
- [Nikolay] prepare pipeline for parsing EPUBs
- [Nikolay] -> run ablations for processed PDFs

Fastext classifier for language:

HIGHLY CONFIDENTIAL - SOURCE CODE Meta_Kadrey_00065306

datasets/books - DONE

Weights (on FAIR cluster):	fair_llm/datasets/tools
GitHub Fasttext code:	
git clone https://github.com/facebook	research,
cd fastText	
make	
pip install.	

Observed OCR parsing artifacts:

Oab5ee32e73a2a455e0cc14894462f69.pdf: In References all references are duplicated

[Nikolay] Loaded 3% of the sci-tech libgen library:

- PDFs: 600GB, 66332 files
- EPUBs: 1.5GB, 781

[Lukas] Smaller model metrics are on par with base model now. Retrained with larger training set.

Model speed -1.8k pages / gpu*hour + 2.2x speed up

dataset/scitech/mmd_small2

[Lukas] 2% error rate per page - i.e. pages are not parsed and skipped

Examples of OCR Parsing

Random books from scitech, pages chosen for diversity

ORIGINAL PARSED WITH OCR

HIGHLY CONFIDENTIAL - SOURCE CODE Meta_Kadrey_00065307

212 Formal integrations and differential equations

5.3 ASYMPTOTIC SOLUTIONS OF O.D.E.S.

5.3.1 Motivation and history

The aim of this part of the book is to describe some recent developments' in the algorithmic methods needed for the "solution" of linear differential equations. Note that here "solution" means "solution in series" We shall only consider equations of the form:

$$a_n(x)(y)^{(n)} + a_{n-1}(x)(y)^{(n-1)} + \cdots + a_0(x)y = 0.$$
 (

where it is always supposed that the o, are polynomials with complex coefficients (we shall discuss this hypothesis later), with no common factor.

Of course, differential equations such as (1) have been the subject of innumerable studies. Ever since the first papers by Gauss in 1812 and those of Kummer (1834), most great mathematicians have worked on solutions to these equations in C. We must mention the papers of Riemann (1857). Weierstrass (1856), Cauchy (1835-1840), before passing on to the findamental work of Fuchs (1865), Frobenius (1873), Poincaré (1881), Birkhoff (1909), to mane only the most important ones. Today these studies have been taken up again by P. Deligne (1976), B. Malgrange (1980) and J.P. Ramis (1981) from the theoretical standpoint.

Why this interest in equations such as (1) Y

There are many answers:

- 1) obvious theoretical interest
- 2) enormous practical interest -- we quote just a few applications of linear differential equations solution by separation of variables of problems with partial derivatives

solution of eigenvalue problems (Sturm-Liouville problems). generation of numerous special functions etc...

What can we hope to contribute to such a branch of mathematics?

5.3 Asymptotic Solutions of O.D.E.S

5.3.1 Motivation and history

The aim of this part of the book is to describe some recent developments at the algorithmic methods needed for the "solution" of linear differential equations. Note that here "solution" means "solution in series". We shall only consider equations of the form:

Footnote * This research is directed by J. Deila Dorn in the Computer Algebra group of the Laboratory LMC at Grenoble, with the help of A. Barkatou, C. Dicrescenzo, A. Hilali, F. Richard-Jung, E. Tournier, A. Wazner, H. Zeili-Naud. The work is carried out in close collaboration with D. Duval, currently at the University of Limoges. with the University of Strasbourg (J.P. Ramis, J. Thoman), and with the Fourier Institute in Grenoble (B.

$$a_n(x)(y)^{(n)} + a_{n-1}(x)(y)^{(n-1)} + \cdots + a_0(x)y = 0$$

where it is always supposed that the a_i are polynomials with complex coefficients (we shall discuss this hypothesis later), with no common factor.

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Why this interest in equations such as (1)? There are many answers:

- Lobytous theoretical interest.
- 2 enormous practical interest we quote just a few applications of linear differential equations solution by separation of variables of problems with partial derivatives solution of eigenvalue problems (Sturm-Liouville problems), generation of numerous special functions etc.

What can we hope to contribute to such a branch of mathematics?

Note: Footnotes are placed after the paragraph

^{*} This research is directed by J. Della Dora in the Computer Algebra group of the Laboratory LMC at Grenoble, with the help of A. Barkatou, C. Dicrescenzo, A. Hilali, F. Richard-Jung, E. Tournier, A. Wazner, H. Zejh-Najid. The work is curried out in close collaboration with D. Duval, currently at the University of Limoges, with the University of Strasbonry (J.P. Ramis, J. Thoman), and with the Fourier Institute in Grenoble. (B. Malgrange).

60 G. Voth

following relationship holds

$$\lim_{T\to\infty} \frac{1}{T} \int_{-T}^{T} dt A_c(t) = (A) \qquad (54)$$

where

$$A_c(t) = \text{Tr} \left\{ \hat{\delta}_c(x_c(t), p_c(t)) \hat{A} \right\} . \qquad (55)$$

This property may not be possessed by many other approximate methods based on, e.g., mean field or semiclassical approaches. Also, in low dimensional systems, the above property is not true for CMD, so to apply CMD to such systems is not consistent with spirit of the method (though perhaps still useful for testing purposes).

On the negative side, the exact time dependent centroid Hamiltonian in Eq. (44) is a constant of motion and the CMD method does not satisfy this condition in general except for quadratic potentials.

V. SOME APPLICATIONS OF CENTROID MOLECULAR DYNAMICS

There has been extensive development of algorithms for carrying out CMD simulations in realistic systems, terms as well as a number of non-trivial applications of the methodology (see, e.g., Ref. 17). In this section, a few illustrative applications will be described. The interested reader is referred to the above citations for more details on CMD algorithms and applications.

V.1 STUDIES ON SIMPLE SYSTEMS

Tests of CMD on simple one-dimensional systems can be carried out by calculating the symmetrized position correlation function:

$$C_{xx}(t) = \frac{1}{2} Tr \left\{ e^{-\beta \hat{H}} \left(\hat{x} e^{t\hat{H}t/\hbar} \hat{x} e^{-t\hat{H}t/\hbar} + e^{t\hat{H}t/\hbar} \hat{x} e^{-t\hat{H}t/\hbar} \hat{x} \right) / 2 \right\} - (56)$$

In the perspective of the centroid time evolution, this correlation function cannot be calculated directly but is obtained through the following relation between the Fourier transforms:

$$\tilde{C}_{xx}(\omega) = \frac{\beta \hbar \omega}{2} \coth \left(\frac{\beta \hbar \omega}{2} \right) \tilde{C}_{xx}^{*}(\omega)$$
 (57)

where C* (to) is the Fourier transform of the Kubo-transformed position correlation function, " The relationship between the latter function and the exact centroid time correlation function, which is calculated approximately by CMD. was established in Ref. 9 as described earlier.

The centroid distribution function and the effective potential for the CMD simulation can be obtained through the path integral simulation method,10 but

following relationship holds

$$\lim_{T\to\infty} \frac{1}{T} \int_{a}^{T} dt A_{a}(t) = \langle A \rangle \qquad (54)$$

where

$$A_c(t) = Tr\{\delta_c(x_c(t), p_c(t))A\}$$
. (55)

This property may not be possessed by many other approximate methods based on, e.g., mean field or semiclassical approaches. Also, in low dimensional systems, the above property is not true for CMD, so to apply CMD to such systems is not consistent with spirit of the method (though perhaps still useful for testing purpose On the negative side, the exact time dependent centroid Hamiltonian in Eq. (44) is a constant of motion and the CMD method does not satisfy this condition in general except for quadratic potentials.

5 Some applications of centroid molecular dynamics

There has been extensive development of algorithms for carrying out CMD simulations in contistic systems [18, 27 28] as well as a number of non-trivial applications of the methodology (see, e.g., Ref. 17). In this section, a few illustrative applications will be described. The interested reader in referred to the above citations for more details: on CMD algorithms and applications

Studies on simple systems

Tests of CMD on simple one-dimensional systems can be carried out by calculating the symmetrized position correlation function:

$$C_{xx}(t) = \frac{1}{Z} Tr \left\{ e^{-\beta A} \left(\delta e^{iAt/\hbar} \delta e^{-iAt/\hbar} + e^{iAt/\hbar} \delta e^{-iAt/\hbar} \delta \right) / 2 \right\}$$
(56)

 $C_{\rm sc}(t) = \frac{1}{Z} Tr \Big\{ e^{-\beta A} \Big(b e^{4At/\hbar} b e^{-At/\hbar} b e^{-At/\hbar} b e^{-At/\hbar} b e^{-At/\hbar} b \Big\} / 2 \Big\} .$ In the perspective of the control fame evolution, this correlation function cannot be calculated directly but is obtained through the following relation between the Fourier transforms

$$\hat{C}_{zz}(\omega) = \frac{\partial \hbar \omega}{2} \coth \left(\frac{\partial \hbar \omega}{2} \right) \hat{C}_{zz}^*(\omega)$$
 (57)

where $\hat{C}_{xx}^{\prime}(\omega)$ is the Fourier transform of the Kubo-transformed position correlation function [15, 25]. The relationship between the latter function and the exact centroid time correlation function, which is calculated approximately by CMD, was established in Ref. 9 as described earlier.

The centroid distribution function and the effective potential for the CMD simulation can be obtained through the path integral simulation method [5, 6], but

Note: In some cases the equation number is added, but not always. We can choose to remove all equation tags.

- · internal nodes representing chemical reaction functions.
- internal nodes representing selector functions that select the reaction's first versus the reaction's second (if any) product.
- external points (leaves) representing substances that are consumed and produced by a reaction.
- · external points representing enzymes that catalyze a reaction, and
- · external points representing numerical constants (reaction rates)

Each program tree in the population is a composition of functions from the problem's function set and terminals from the problem's terminal set.

Repertoire of Functions

There are four chemical reaction functions and two selector functions.

The first argument of each chemical reaction (CR) function identifies the enzyme that catalyzes the reaction. The second argument specifies the reaction's rate. In addition, there are two, three, or four arguments specifying the substrate(s) and product(s) of the reaction. Table 5.1 shows the number of substrate(s) and product(s) and overall arity for each of the four chemical reaction functions. The runs in this chapter use a first-order and second-order rate law.

Table 5.1 From chemical reaction functions

Function	Substrates	Products	Army
CK_LJ	1	1	4
CR.1.2	1	2	5
CR.2.1	7	T.	3.
CR22	2	2	0

Each function returns a list composed of the reaction's one or two products. The one-argument FIRST function returns the first of the one or two products produced by the function designated by its argument. SECOND function returns the second of the two products (or, the first product, if the reaction produces only one product).

Repertoire of Terminals

Some terminals represent substances (input substances, intermediate substances created by reactions, or output substances). Other terminals represent the enzymes that catalyze the chemical roactions. Still other terminals represent journerical constants for the rate of the roactions. · internal nodes representing chemical reaction functions

- internal nodes representing selector functions that select the reaction's first versus the reaction's second (if any) product.
- · external points (leaves) representing substances that are consumed and produced by a reaction.
- · external points representing enzymes that catalyze a reaction, and
- external points representing numerical constants (reaction rates).

Each program tree in the population is a composition of functions from the problem's function set and terminals from the problem's terminal set

5.1.1 Repertoire of Functions

There are four chemical reaction functions and two selector functions.

The first argument of each chemical reaction (CR) function identifies the enzyme that catalyzes the reaction. The second argument specifies the reaction's cate. In addition, there are two, three, or four arguments specifying the substrates(s) and product(s) of the reaction. Table 5.1 shows the number of substrate(s) and product(s) and overall arry for each of the four chemical reaction functions. The runs in this chapter use a first-order and second-order rate law.

Each function returns a list composed of the reaction's one or two products. The one-argument FIRST function return the first of the one or two products produced by the function designated by its argument. The one-argument SEC OND function returns the second of the two products (or, the first product. If the reaction produces only one product.)

5.1.2 Repertoire of Terminals

Some terminals represent substances (input substances, intermediate substances created by reactions, or output substances). Other terminals represent the eazymes that vatalyze the chemical reactions. Still other terminals represent numerical constant for the rate of the reactions.

Function	Substrates	Products	Arity
CR_t_t	1	1	4
CR LZ	1	2	3
CR_2_1	2	1	5
CR 2 2	2	7	6

Table > 1. Four chemical reaction functions

Note: Sometimes the model hallucinates subsection numbers (here from the table label) due to training data impurity. We can choose to filter out all section numbering. Also, tables and figure captions will always be placed at the end of the page

(iii) Automated Severie Engineering of Metabolic Pathware by Genetic Programming

00	OF	2	$\overline{}$	α
02.	UE	2.6	u	۷:

[Lukas] Smaller decoder model has a 2x greater conversion speed. Metrics are slightly worse but parsing samples look similar PDF parsing samples smaller model: dataset/scitech/mmd_small

28.04.2023

[Lukas] Parsed with OCR library 70 books (29,488 pages total), it took 18 hours on 2 GPUs -> 2 books / gpu*hour -> ~800 pages / gpu*hour

- sci-tech: 3,274,071 books * 51% EN * 65% PDFs = 1M books = 260M pages
 260M pages / (500 pages / hour*gpu) = 500k GPU+hours -> so with 1000 GPUs it will take 500 hours (20 days)
 \$25 / GPU day -> 1000*20*\$25 = \$0.5M (VS \$16M
- sci-mag: 72,624,976 articles * 50% EN * 6 pages = 220M pages
 220M pages / (500 pages / hour*gpu) = 440k GPU*hours -> 18 days with 1000 GPUs

PDF parsing samples:

26.04.2023

There is a sample of downloaded libgen documents on fair cluster (totals taken from here):

- fiction: 126GB(2% of total 5.6TB)
- scitech: 9.3TB (16% of total 59.4TB)
- scimag: 397GB (0.5% of total 80.6TB)

Fair cluster -> Python Lib torrent (list of magnet links) 50 torrents -> 2 days

Some processed samples from scitech on fair cluster:

data_v2/datasets/books/data/scitech_pdf/

scitech processed PDFs: 63GB

24.04.2023

Reading metadata from the MySQL dumps: http://libgen.rs/dbdumps/. There are 3 category of content:

- Fiction: fiction.rar ->1,607,593 unique records (title&author)
- Scitech: libgen.rar -> 3,274,071 unique records (title&author)
- Scimag: scimag.sql.gz -> TBD

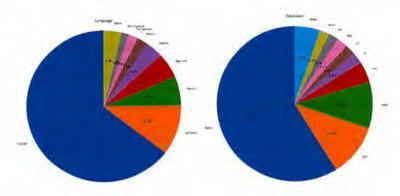
Findings:

- Each DB dump contains metadata (table: fiction), book description (table: fiction_description) and hashes (table: fiction_hashes)
- Hashes table (fiction_hashes) provides the hashes to download files using torrents or IPFS (InterPlanetary File System file sharing peer-to-peer network):
 - Torrent (using BitTorrent Info Hash: 'btih'): magnet/?εt=urn:btih: YOUR_BT_HASH -> paste this link into qBittorrent or μTorrent, or Transmission.
 - IPFS downloads (using 'ipfs_cid'): https://ipfs.io/ipfs/YOUR_IPFS_CID
 - Other columns: 'md5', 'crc32', 'edonkey', 'aich', 'sha1', 'tth', 'btih', 'sha256', 'ipfs_cid'
- LibGen is a different project and database from Sci-Hub. The sci-rech section of LibGen focuses on scientific and technical books, while the sci-mag section
 provides access to scientific and academic journal articles, which is the primary focus of Sci-Hub.

Fiction

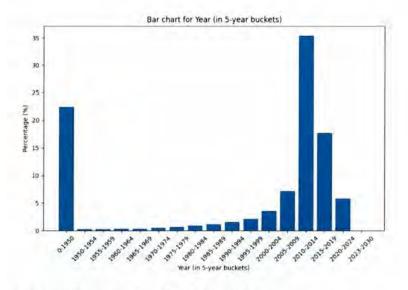
Tables: fiction, fiction_description, fiction_hashes

- fiction table num_records: 2,693,056
- columns: ['ID', 'MD5', 'Title', 'Author', 'Series', 'Edition', 'Language', 'Year', 'Publisher', 'Identifier', 'GooglebookID', 'ASIN', 'Coverurl', 'Extension', 'Filesize', 'Library', 'Issue', 'Locator', 'Commentary', 'Generic', 'Visible', 'TimeAdded', 'TimeLastModified']
- English: 65% | German: 11% | French: 6%
- Epub: 59% | PDF: 11% | mobi: 10%
- 0.5M books without a year



HIGHLY CONFIDENTIAL - SOURCE CODE

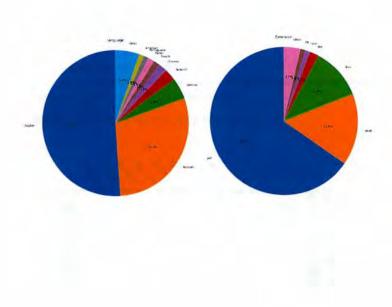
Meta_Kadrey_00065313



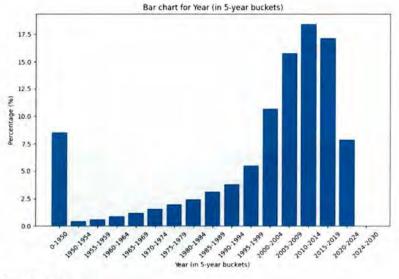
Sci-tech (libgen - main sci-tech collection)

Description: https://wiki.mhut.org/catalog:database

- Tables: updated (main metadata table), updated_edited, description, description_edited, hashes, topics
- updated table num_records: 3,706,772
- English: 51% | Russian 29% | German: 5%
- Epub: 16% | PDF: 65% | djvu: 11%

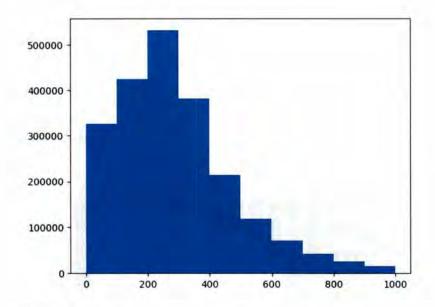


HIGHLY CONFIDENTIAL - SOURCE CODE Meta_Kadrey_00065314.00001



Pages distribution

HIGHLY CONFIDENTIAL - SOURCE CODE

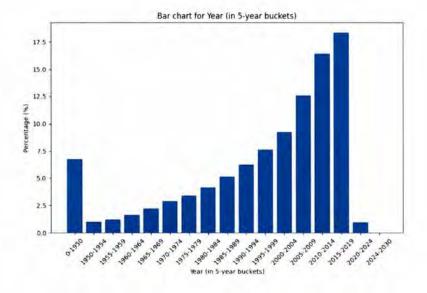


Sci-mag

- Tables: scimag, publishers, magazines, error_report
- fiction table num_records: 2,693,056
- columns: ['ID', 'MD5', 'Title', 'Author', 'Series', 'Edition', 'Language', 'Year', 'Publisher', 'Identifier', 'GooglebookID', 'ASIN', 'Coverurl', 'Extension', 'Filesize', 'Library', 'Issue', 'Locator', 'Commentary', 'Generic', 'Visible', 'TimeAdded', 'TimeLastModified']

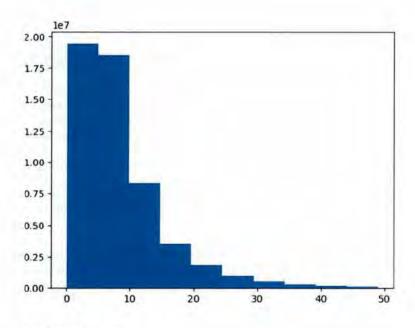
- English: 65% | German: 11% | French: 6%
- Epub: 59% | PDF: 11% | mobi: 10%
- Scientific articles in this dump are before May 2020





Pages distribution

HIGHLY CONFIDENTIAL - SOURCE CODE Meta_Kadrey_00065314.00005



21.04.2023

Key takeaways:

- Only 3% of books from in epub format are in LibGen (out of 1000 sample).

- http://libgen.rs/dbdumps/: libgen metadata dumps [loading this 1.1 GB fiction.rar file takes 10 hours 👩 could use Folx to download in multiple threads]
- http://libgen.rs/scimag/repository_torrent/: torrent files for scimag
- https://phillm.net; some indexer of torrent seeds
- https://ipfs.io/ipfs/bafkreibjbw2czkimwt5q7yeu3wko3a2fuw6q4km7rwo2wweirc6oejmokm: candidat for metadata DB dump

19.04.2023

We need to come up with a reliable book matching algorithm. There are many books with similar titles (ex. C/C++), so we need to account for authors' matches as well (at least partial authors match). The matching algorithm used checks for the exact Title match and at least one of the authors match.

Results:

- Up to 90% of books are present in LibGen for
- The books in LibGen are in djvu/epub/pdf format, so the parsing quality would be worse compared to getting the books from publishers directly. However epub is almost the same as HTML it's a ZIP archive containing a collection of HTML, CSS. So we can extract text without losing quality from it.
- The books in LibGen often have a previous edition (compared to the ones in

Caveats:

- Matching algorithm is not perfect as well as the LibGen API (so up to 5% false negatives could be present)
- Sampling from all available titles is not perfect (pseudo random), especially for the problem is that we don't have the full list of titles for either of the publishers, so we need to scrape their web-pages. For that I sampled random beginning letters and random pages from around certain alphabetic characters

Publisher	Method	Match (%)
-----------	--------	-----------

Titles&Authors from sampled web-scraping	90% (sample=1000)
Manual check	88% (sample=25)
Titles&Authors from sampled web-scraping	68% (sample=1000)
Manual check	76% (sample=25)

Code:

- Notebook LibGen VS
- Web-Scraping
- Web-Scraping
- Utils for Web-Schttps://www.internalfb.com/
- Quick Manual Check: LibGen VS
- VS
- Quick manual check

18.04.2023

Motivation: Collect available book titles and authors from Cobservations:

1. We don't have a full list of titles for the collection of th

The problem in this approach is that Springer has 300k book titles and each book has it's own page with details that we need. A lot of requests to be made (possible DDOS)

i. https://link.com/books/g/1
ii. https://www.com/en-us/

b. Use APIs

Only has APIs for accessing their resources, but it is limited to 100 result per subject. So you first get sample DOI for each category in then request details for these DOI. In total you can get 140 books meta (out of 300k) and 2k articles (which we are less interested).

- c. Manual check on their website and randomly checking 25 books
- One should be careful with doing too many requests to web-resources I got blocked by LibGen after 1k requests in a few minutes (after I tired multithreading+multiprocessing together).

a

3. LibGen API can be missing results (ex. I can find a title manually, but the API doesn't return anything), most likely the API is using a different database. But this is <5% of cases.

Results:

- 1. Prepared scripts for web-scraping
- 2. Prepared scripts for checking the books in LibGen

Appendix

Links:

- Libgen API: https://pypi.org/project/libgen-api/

- Libgen Search: https://libgen.li
- Sample of documents on fair_cluster:
- Some description of the project: https://news.ycombinator.com/item?id=21692841
- Libgen Books Metadata: http://libgen.rs/dbdumps/
 - https://link.
- https://www.pearson.com/en-us/

Plan:

- 1. [in parallel] Find out where to get the dump of the datasets (scitech, fiction and scimag):
 - a. Taking metadata from here: http://libgen.rs/dbdumps/
 - SQL search: It seems that they have the database dumps which I assume are behind the API. It would be much faster to create an SQL database (I assume they use mysql or postgres) which we can setup locally. Then querying is fast.
 - ii. Embedding/Elastic search: It might make sense to have some embedding search using fastext embeddings. Encode everything 100M records with fastext(title), fastext(author), fastext(abstract??). If presented, it would be relatively cheap to search. Then match the concat(ft_title, ft_author, ft_abstract). BoW with wparse char 3-grams should work too.
 - b. Run some high-level stats: share of epub/pdf, share of EN, total count of books, etc...
 - c. Decide on where to store the files: aprox. ~120TB * 30% (english & PDF/EPUB) = ~40TB
 - d. Load the dataset (we probably need filtered data: English and only PDF+EPUB format). Should we load to Meta's Manifold bucket instead of S3?
- 2. [in parallel] Compare quality of text extraction from LibGen VS
 - a. Load samples of pdfs/epubs from the libgen website https://libgen.is/, same samples as from
 - b. Check % of samples in libgen epub only format
 - c. Parse epub with Marie-Anne Lachaux's html script
 - d. Parse pdf with Lukas's OCR script, record the speed of parsing to further estimate the GPU requirements
 - e. Compare quality VS data (original pdfs)
- 3. [in parallel] Check what books we have in CC (as per Todor Mihaylov's suggestion)

- a. Check quality/format
- b. Check intersection with titles / LibGen titles (Nikolay Bashlykov to provide code for checking titles using libgen-api)
- 4. [once data loaded] Filtering & Preprocessing
 - a. Filtering rules
 - b. Run ablations

To Discuss:

- Can we load libgen data using Meta IP ranges? Or should we use some vpn?
 Redacted Privilege (to check with Marie-Anne and Guillaume)

 Can we load this data to S3? Or use Meta's Manifold solution? You can load data to RSC from Manifold straightaway and Redacted Privilege

 Redacted Privilege Is there any preference to use manifold from a tech perspective?
- [Todor] No, because we need to process it on AWS/fairspark.
 Is there any overlap between the big dump of cc pdfs and libgen pdfs?
 - [Mel] asking so we don't duplicate processing/can prioritize a bit. Maybe easy version is hashing
 - Don't know yet, can try hashing/comparing titles from metadata
- How clean can we get scientific PDFs? Do we still want to buy
- How long will it take for a first pass of data to be ready?
 - Should we include in v3 or is this too not trending to higher quality models based on our ablations and/or do we feel it is too risky to change our data mix?
 - Should we hold 150B training for this?
 - [Nikolay/Peter] May 17th might for the whole set would be right; common crawl PDFs seem more doable by then. Just the epub may be possible but need tighter estimates on downloading time (possibly bottlenecked on the p2p network)
 - [Mel] Let's try to batch downloading and processing so we can get some of the data in weeks instead of all of the data in months.
- How much of the datastet is Pdfs? What portion can we use pdf extract for vs need to OCR? how many GPUs is it going to take to OCR the parts of the dataset that can't be pdf extracted for how long (good to know this ASAP)? -> TBD

- [Peter] 3M books, OCR takes 10 seconds per page/20 mins per book ⇒ 1M GPU hours, 3 weeks for 3K GPUs.
- [Todor] estimate above sounds too high; output might be bad quality

Still to answer: tighter timeline estimation for first batch of data,

- [Nikolay] estimation for the first batch TBD 29.04 -> run ablation on the first chunk by 12.05
- [Nikolay] re I don't think we need to proceed with
 - overlaps with up to 90% of content in LibGen
 - Quality in LibGen seems to be very high (from a sampled check) for the Sci-tech collection (similar to
 - LibGen is at least 6 times as large as 1.4M books (sci-tech EN books in PDF&Epub) VS 212k (EN books in LibGen VS 3M EN articles in
- [Nikolay] re GPUs needed: with Lukas's estimates on PDF parsing we would need optimally 2k GPUs for OCR parsing to complete in sci-tech in 10 days. And additional 10 days for sci-mag (with less priority). We would need these resources from:
 - fiction: 0
 - sci-tech: 500k GPU*hours
 - sci-mag: 440k GPU*hours (lower priority)
- [Nikolay UPD 28.04] we were able to accelerate the OCR parsing by over 2.5x, so the required GPU*hours would be 2.5x less. We are still analyzing the parsing quality tradeoffs, as this is a smaller model.

Commented [9]: Remaining things to answer